

Fun with High Throughput Toxicokinetics

Webinar Presentation to CalEPA Office of Environmental Health Hazard Assessment

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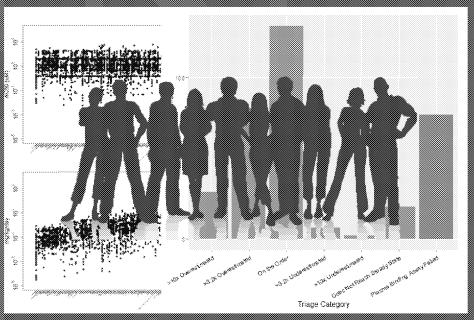


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Introduction

- Toxicokinetics (TK) provides a bridge between toxicity and exposure assessment by predicting tissue concentrations due to exposure
 - However traditional TK methods are resource intensive
- Relatively high throughput TK (HTTK) methods have been used by the pharmaceutical industry to determine range of efficacious doses and to prospectively evaluate success of planned clinical trials (Jamei, et al., 2009; Wang, 2010)
 - A key application of HTTK has been "reverse dosimetry" (also called Reverse TK or RTK)
 - RTK can approximately convert in vitro HTS results to daily doses needed to produce similar levels in a human for comparison to exposure data (starting off with Rotroff, et al., 2010)
- A new EPA open source R package ("httk") is freely available on CRAN allows RTK and other statistical analyses of 553 chemicals (more coming)

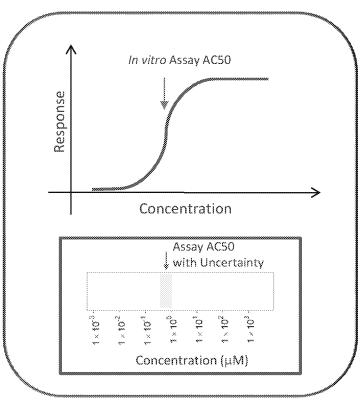
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High-Throughput Bioactivity

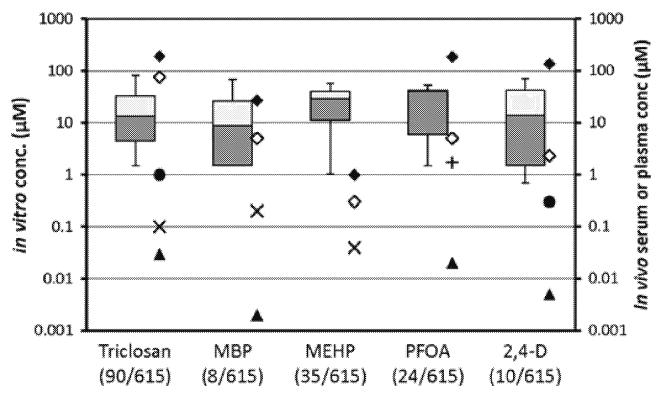
- ▼ Tox21: Examining >8,000 chemicals using ~50 assays intended to identify interactions with biological pathways (Schmidt, 2009)
- **ToxCast**: For a subset (>2000) of Tox21 chemicals ran >1100 additional assays (Judson *et al.*, 2010)
- Most assays conducted in dose-response format (identify 50% activity concentration AC50 and efficacy if data described by a Hill function, Filer et al., 2016)
- All data is public: http://comptox.epa.gov/





in vitro - in vivo Concordance





Aylward and Hays (2011)
Journal of Applied Toxicology **31** 741-751

- estimated or measured average concentrations associated with the LOAEL in animal studies
- NOAEL in animal studies
- Humans with chronic exposure reference values (solid circles)
- X Volunteers using products containing the chemical
- + Biomonitored occupational populations
- General populations

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High Throughput Risk Prioritization

mg/kg BW/day

- High throughput risk prioritization relies on three components:
 - high throughput hazard characterization
 - high throughput exposure forecasts
 - high throughput toxicokinetics (i.e., dosimetry)
- While advances have been made in toxicity and Potential Exposure exposure screening, TK methods applicable to 100s of chemicals are needed

Potential Hazard from in vitro with Reverse Toxicokinetics from ExpoCast Higher

Risk

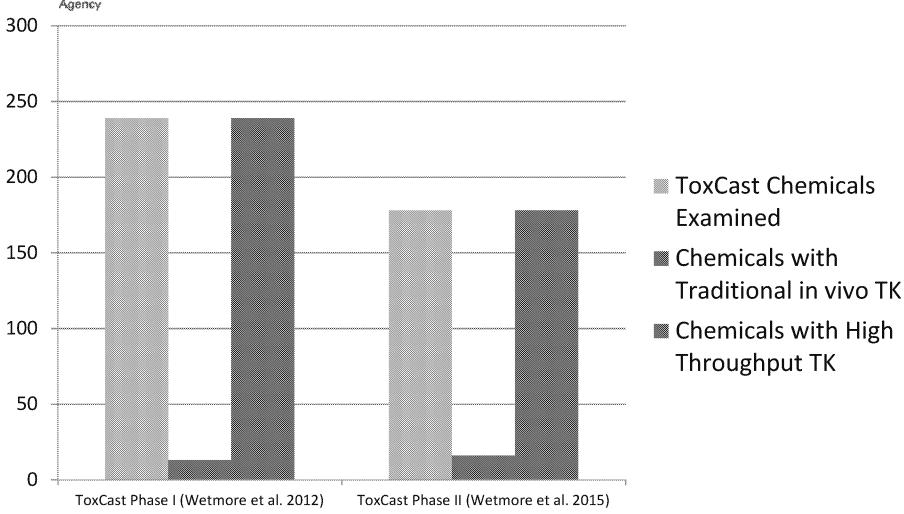
see Wetmore et al. (2015)

Medium Risk

Risk



The Need for *In Vitro*Toxicokinetics



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 Studies like Wetmore et al. (2012, 2015), address the need for TK data using in vitro methods



In Vitro - In Vivo Extrapolation (IVIVE)

Definition:

IVIVE is the utilization of in vitro experimental data to predict phenomena in vivo

- IVIVE-PK/TK (Pharmacokinetics/Toxicokinetics):
 - Fate of molecules/chemicals in body
 - Considers absorption, distribution, metabolism, excretion (ADME)
 - Uses empirical PK and physiologically-based (PBPK) modeling
- IVIVE-PD/TD (Pharmacodynamics/Toxicodynamics):
 - Effect of molecules/chemicals at biological target in vivo
 - Assay design/selection important
 - Perturbation as adverse/therapeutic effect, reversible/irreversible
- Both contribute to predict in vivo effects

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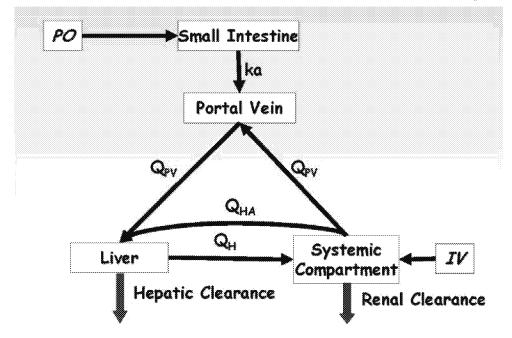
High Throughput Toxicokinetics (HTTK)

Jamei et al. (2009)

Minimal Model: Lumped Single Distribution Volume

\$ 2000 (De tou getoure)

- In vitro plasma protein binding and metabolic clearance assays allow approximate hepatic and renal clearances to be calculated
- At steady state this allows conversion from concentration to administered dose
- 100% bioavailability assumed

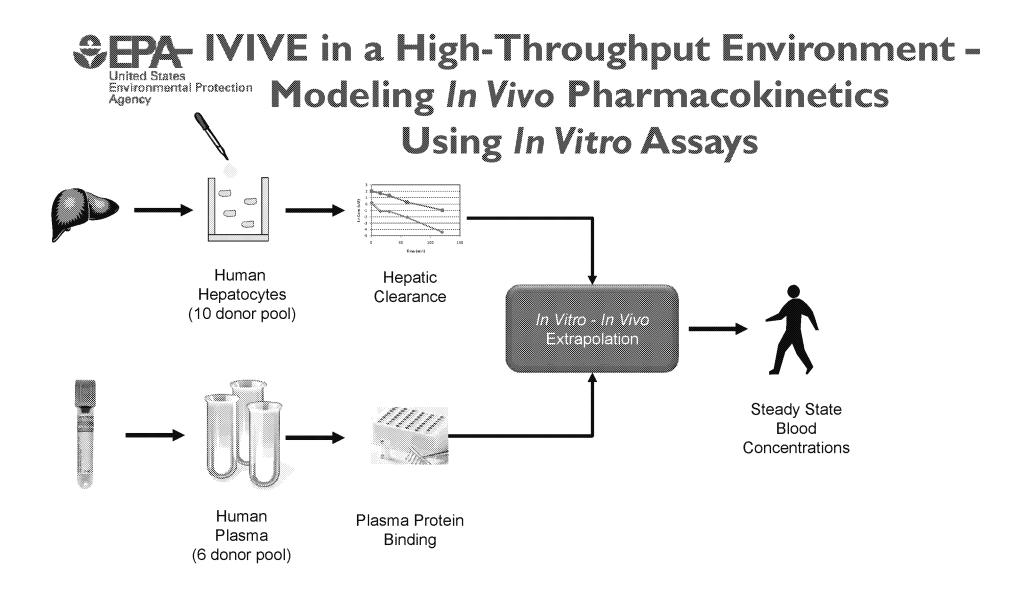


$$C_{ss} = \frac{\text{oral dose rate}}{\left(\text{GFR * F}_{ub}\right) + \left(Q_1 * F_{ub} * \frac{Cl_{int}}{Q_1 + F_{ub} * Cl_{int}}\right)}$$

Sum of hepatic and renal clearance (mg/kg/day)

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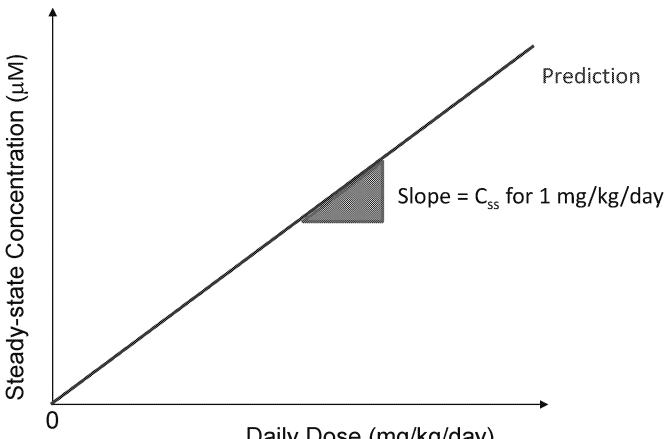
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Slide from Barbara Wetmore



Steady-State is Linear with Dose



$$C_{ss} = \frac{\text{oral dose rate}}{\left(\text{GFR} * F_{ub}\right) + \left(Q_1 * F_{ub} * \frac{Cl_{int}}{Q_1 + F_{ub} * Cl_{int}}\right)}$$

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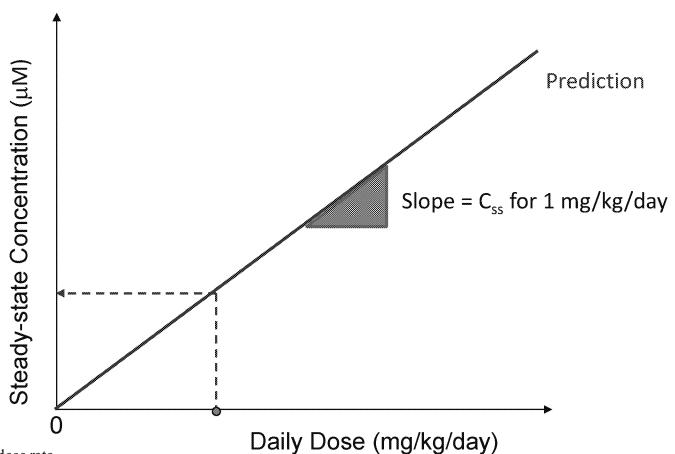
Wetmore et al. (2012)

Daily Dose (mg/kg/day)

Can calculate predicted steady-state concentration (C_{ss}) for a 1 mg/kg/day dose and multiply to get concentrations for other doses



Steady-State is Linear with Dose



$$C_{ss} = \frac{\text{oral dose rate}}{\left(\text{GFR} * F_{ub}\right) + \left(Q_1 * F_{ub} * \frac{Cl_{int}}{Q_1 + F_{ub} * Cl_{int}}\right)}$$

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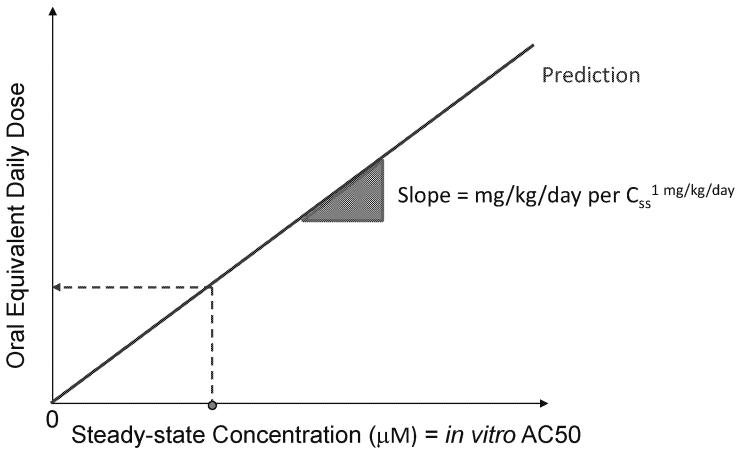
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Wetmore et al. (2012)

Can calculate predicted steady-state concentration (C_{ss}) for a 1 mg/kg/day dose and multiply to get concentrations for other doses



HTTK Allows Steady-State In Vitro-In Vivo Extrapolation (IVIVE)



- Swap the axes (this is the "reverse" part of reverse dosimetry)
- \sim Can divide bioactive concentration by C_{ss} for for a 1 mg/kg/day dose to get oral equivalent dose

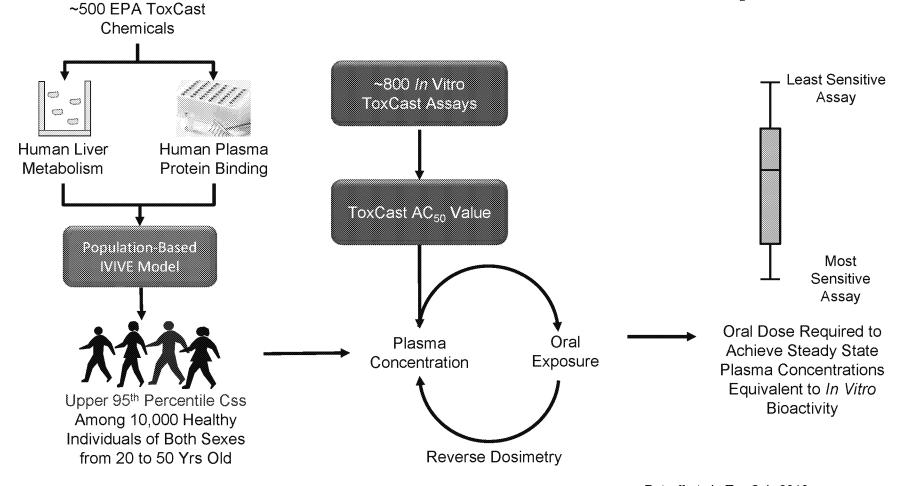
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Wetmore et al. (2012)



Integrating Human Dosimetry and Exposure with ToxCast In Vitro Assays



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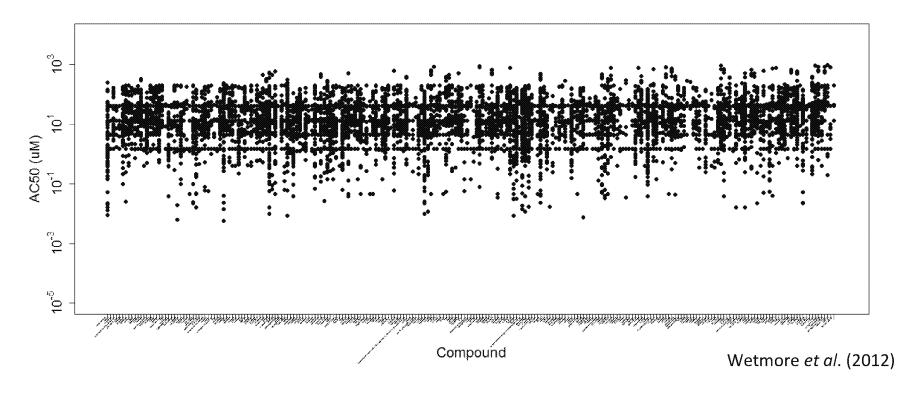
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Rotroff et al., Tox Sci., 2010 Wetmore et al., Tox Sci., 2012 Wetmore et al., Tox Sci, 2015

Slide from Barbara Wetmore



ToxCast in vitro Bioactive Concentrations

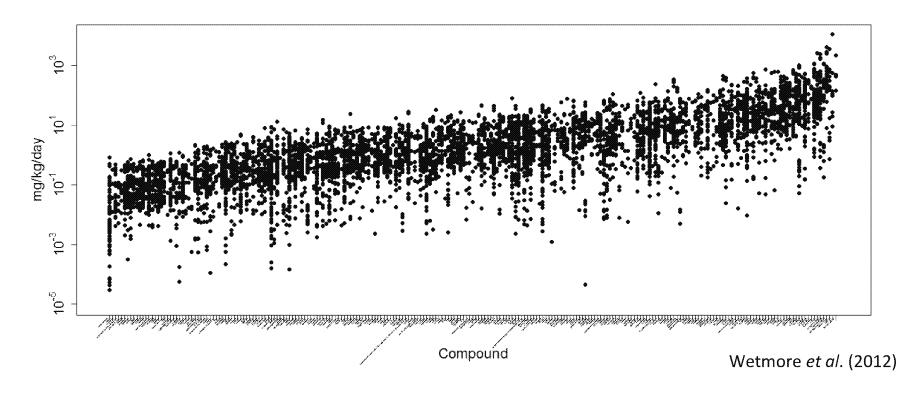


It appears harder to prioritize on bioactive in vitro concentration without in vivo context

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HTTK Oral Equivalents



Translation from in vitro to steady-state oral equivalent doses allow greater discrimination between effective chemical potencies

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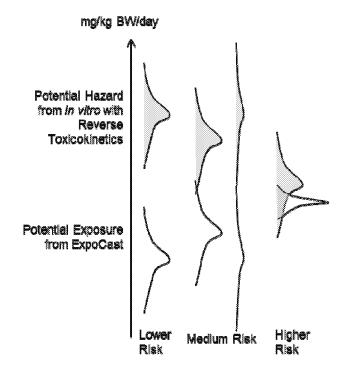
Activity-Exposure Ratio

(Wetmore et al. 2012, 2014, 2015)

$$AER = \frac{Oral Equiv. Dose}{Estimated exposure}$$

AER <=1: Exposure potentially high enough to cause bioactivity

AER >> 1: Exposure less likely to be high enough to cause bioactivity



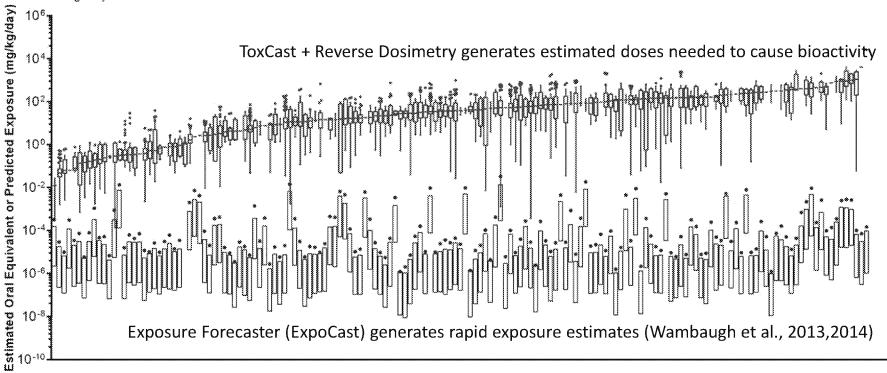
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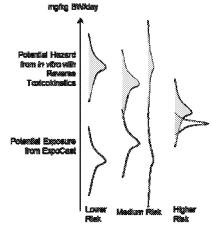
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Slide from Caroline Ring



Incorporating Dosimetry-Adjusted ToxCast Bioactivity Data with Exposure





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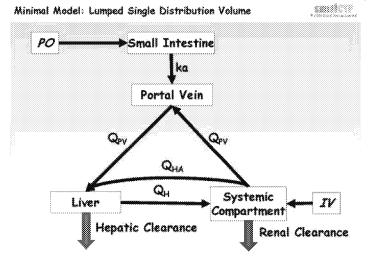
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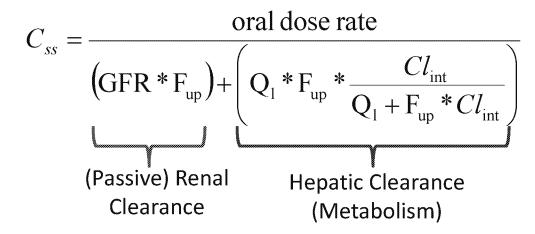
Wetmore et al., Tox. Sci, 2015



Variability in this Steady-State TK Model

Jamei et al. (2009)



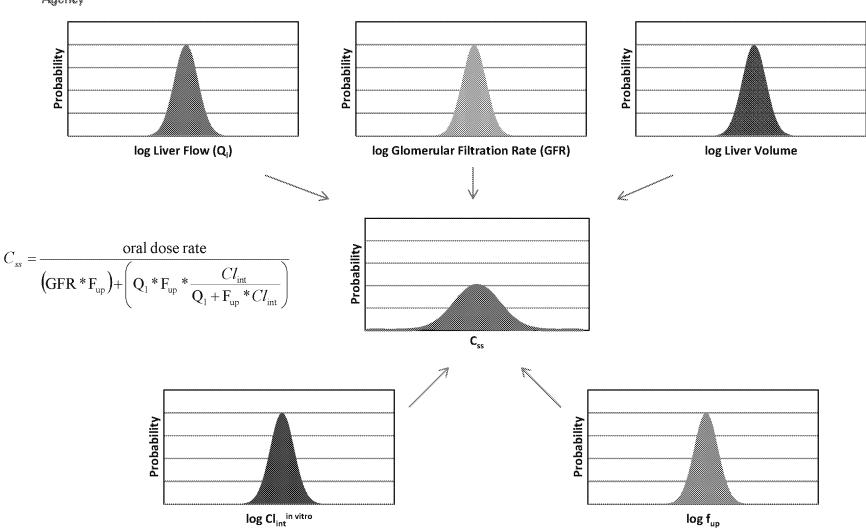


- In vitro clearance (μL/min/10⁶ hepatocytes) is scaled to a whole organ clearance using the density of hepatocytes per gram of liver and the volume of the liver (which varies between individuals)
- Glomerular filtration rate (GFR) and blood flow to the liver (Q_I) both vary from individual to individual
- Further assume that measured HTTK parameters have 30% coefficient of variation

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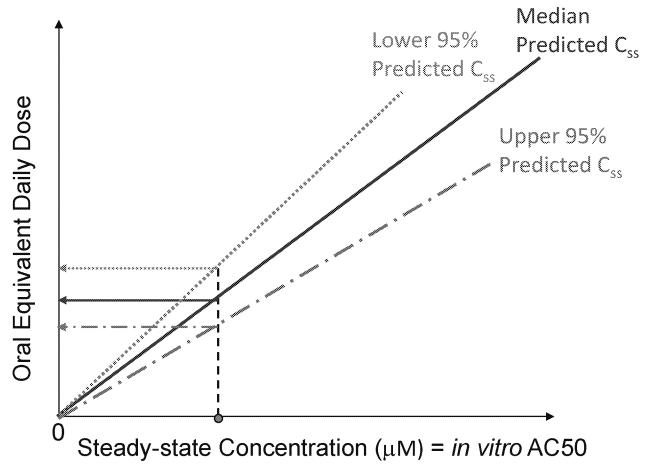
Monte Carlo (MC) Approach to Variability



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Steady-State In Vitro-In Vivo Extrapolation (IVIVE)



The higher the predicted C_{ss} , the lower the oral equivalent dose, so the upper 95% predicted C_{ss} from the MC has a lower oral equivalent dose

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HTTK Limitations (from Ring et al., 2017)

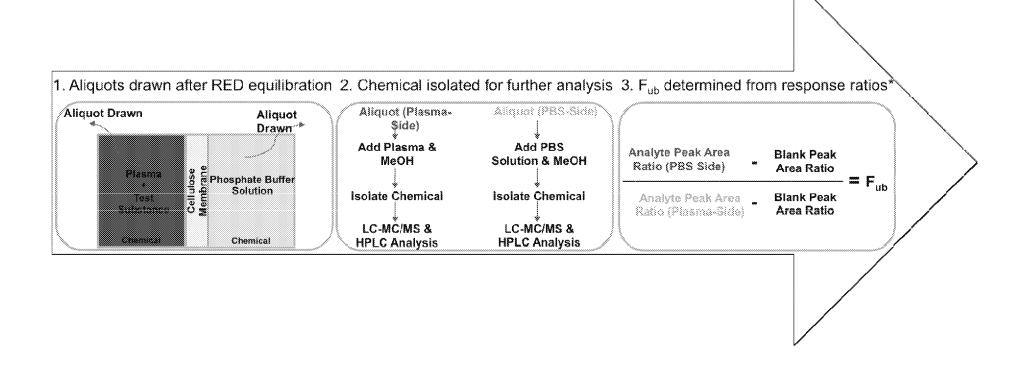
- Oral absorption
 - 100% assumed, but may be very different
 - In silico models not necessarily appropriate for environmental chemicals
- Hepatic Clearance (CL_{int})
 - Ten donor pool in suspension for 2-4 h misses variability and low turnover compounds
 - Isozyme abundances and activity: varies with age, ethnicity (at least) (Yasuda et al. 2008, Howgate et al. 2006, Johnson et al. 2006)
 - Parent chemical depletion only
- Isozyme-specific data & modeling (Wetmore et al. 2014)
 - Isozyme-specific metabolism assays not HT
 - In silico predictions of isozyme-specific metabolism? Not easy!
 - Existing data is mostly for pharmaceuticals
- Plasma binding assay (F_{un})
 - Assay often fails due to analytical chemistry sensitivity (Wetmore et al., 2012)
 - Plasma protein concentration variability (Johnson et al. 2006, Israili et al. 2001)
 - Albumin or AAG binding? (Routledge 1986)

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Plasma Protein Binding Assay is Limited by Analytical Chemistry

Rapid Equilibrium Dialysis (RED) Method: Waters et al. (2008)





Why Build Another PBTK Tool?

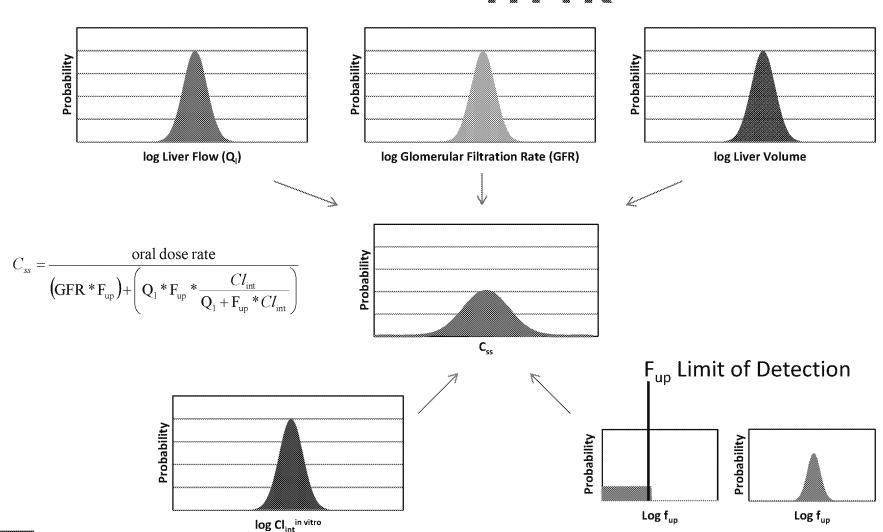
	SimCYP	ADMET Predictor / GastroPlus	MEGen	httk
Maker	SimCYP Consortium / Certara	Simulations Plus	UK Health and Safety Laboratory (Loizou)	US EPA
Availability	License, but inexpensive for research	License, but inexpensive for research	Free: http://xnet.hsl.gov.uk/mege n	Free: CRAN Repository
Population Variability Monte Carlo	Yes	No	No	Yes
Batch Mode	Yes	Yes	No	Yes
Physiological Data	Yes	Yes	Yes	Yes
Chemical-Specific Data Library	Clinical Drugs	No	No	Pharma and ToxCast Compounds: 443 PBTK, +100 steady-state only
Export Function	No	No	Matlab and AcslX	SBML and Jarnac
R Integration	No	No	No	Yes
Easy Reverse Dosimetry	Yes	Yes	No	Yes
Future Proof XML	No	No	Yes	No

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We want to do a statistical analysis (using R) for as many chemicals as possible



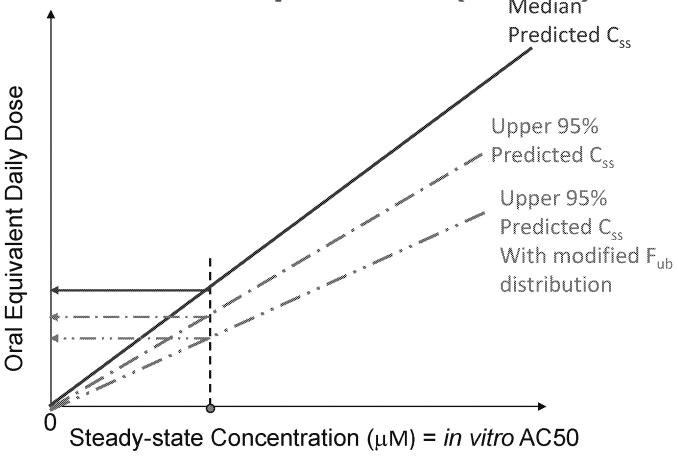
Modified f_{up} Distribution for HTTK



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Steady-State In Vitro-In Vivo Extrapolation (IVIVE) Median

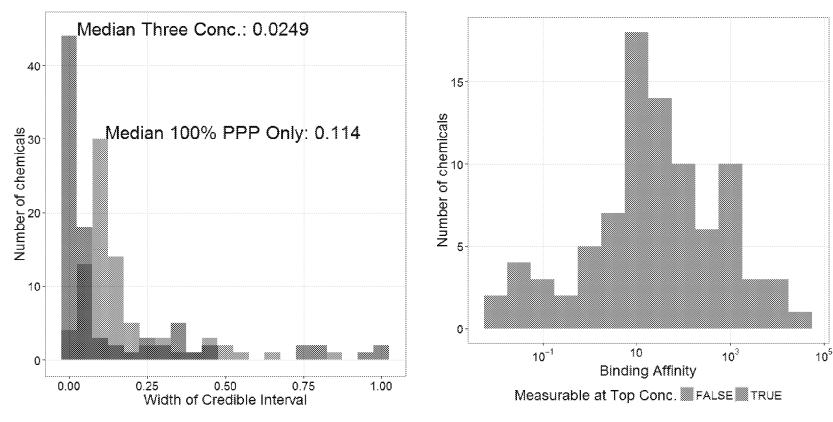


Taking into account the limit of detection issues does not change the median (or lower 95% C_{ss}) but does change the upper C_{ss} , causing lower oral equivalent dose predictions (greater sensitivity)

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Improving Plasma Binding Measurement



- Using a Bayesian analysis via MCMC (in JAGS) to estimate 95% credible intervals
- New protocol uses three plasma protein concentrations (100%, 30%, and 10% of physiologic concentration)
- · Can analyze data jointly using a binding affinity model

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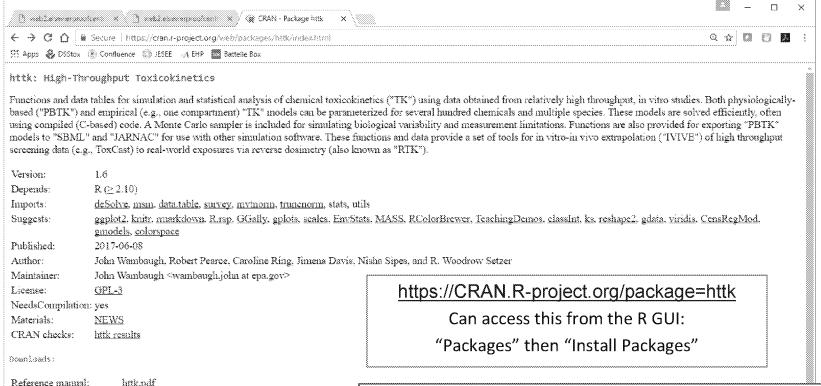


Goals for HTTK

- In order to address greater numbers of chemicals we collect in vitro, high throughput toxicokinetic (HTTK) data
- The goal of HTTK is to provide a human dose context for in vitro concentrations from HTS
 - This allows direct comparisons with exposure
- An R statistical package allows us to evaluate in vitro predictions two ways:
 - We compare *in vitro* predictions and *in vivo* measurements
 - We perform simulation studies to examine key assumptions

R Package "httk"





Creating Partition Coefficient Evaluation Plots Vignettes:

Age distributions

Global sensitivity analysis Global sensitivity analysis plotting Height and weight spline fits and residuals Hematocrit spline fits and residuals

Piotting Css95

Serum creatinine spline fits and residuals

Generating subpopulations

Evaluating HTTK models for subpopulations

Generating Figure 2 Generating Figure 3

Piotting Howgate/Johnson data

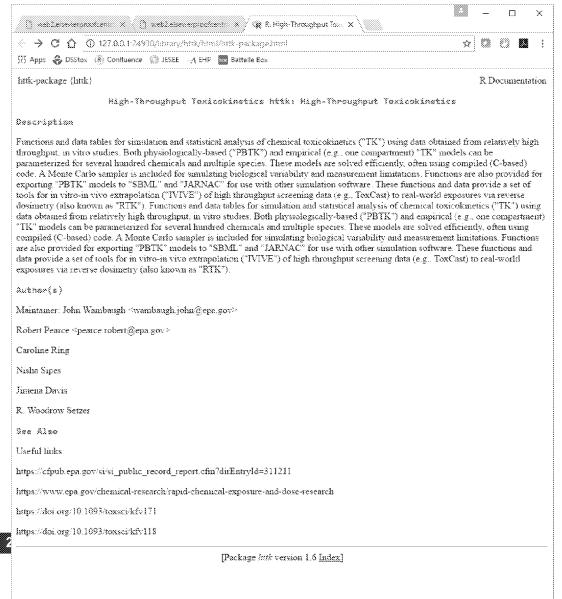
AER piotting

Virtual study populations

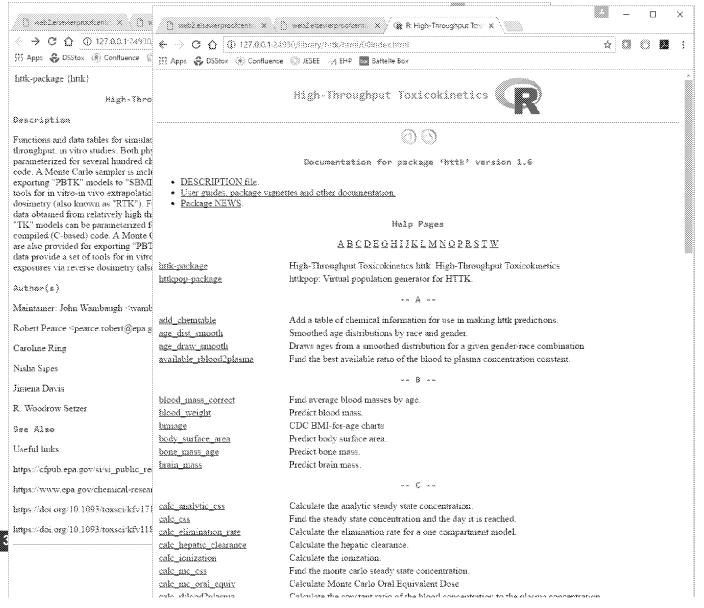
- "httk" R Package for reverse dosimetry and PBTK
- 553 chemicals to date
- 100's of additional chemicals being studied
- Pearce et al. documentation manuscript accepted at Journal of Statistical Software
- Vignettes provide examples of how to use many **functions**

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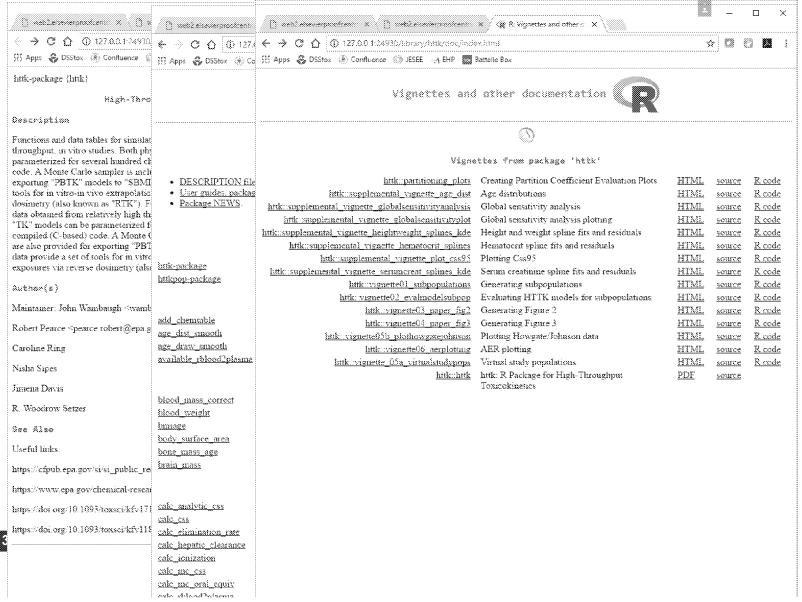




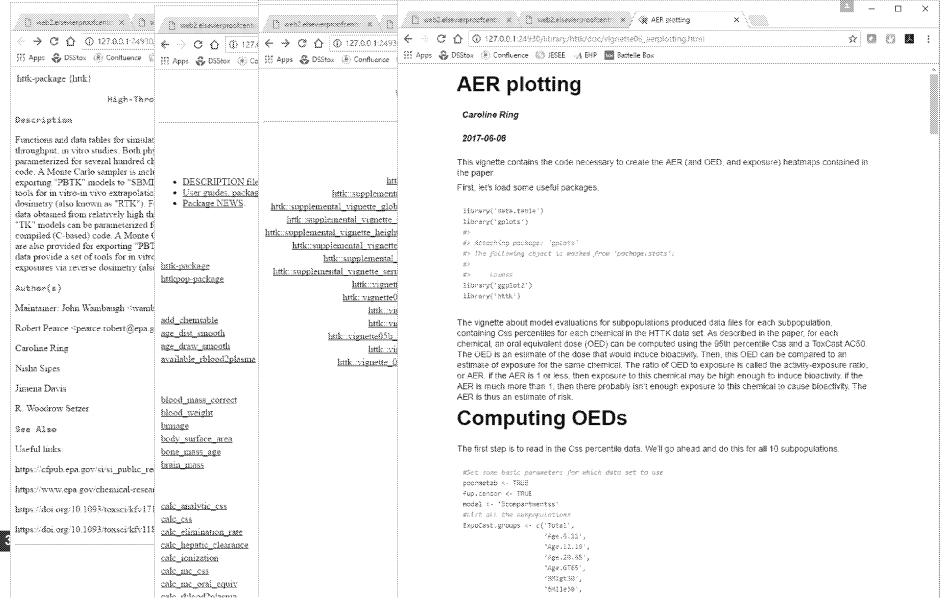














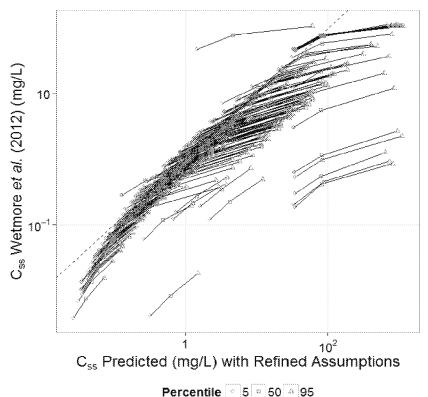
What you can do with R Package "httk"

- Allows, one compartment, two-compartment, three-compartment, and PBTK modeling
- Allows conversion of in vitro concentration to in vivo doses
- Allows prediction of internal tissue concentrations from dose regimen (oral and intravenous)
- A peer-reviewed paper in the Journal of Statistical software provides a how-to guide (Pearce et al., in press)
- You can use the built in chemical library or add more chemical information (examples provided in JSS paper)
- You can load specific (older) versions of the package
- You can use specific demographics in the population simulator (v1.5 and later Ring et al., in press)
 - Gender, age, weight, ethnicity, renal function
- You can control the built in random number generator to reproduce the same random sequence

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Comparison Between httk and SimCYP



- In the Rotroff *et al.* (2010) and Wetmore et al. (2012,2013,2014,2015) papers SimCYP was used to predict distributions of C_{ss} from *in vitro* data
 - We show that "httk" can reproduce the results from those publications for most chemicals using our implementation of Monte Carlo.
- Any one chemical's median and quantiles are connected by a dotted line.
- The RED assay for measuring protein binding fails in some cases because the amount of free chemical is below the limit of detection
 - A default value of 0.5% free was used
 - Now we use random draws from a uniform distribution from 0 to 1%.

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Wambaugh et al. (2015)



Steady State Concentration Examples

library(httk)

```
#Steady-state concentration (uM) for 1 mg/kg/day for 0.95 quantile for human for Acetochlor (published value):
calc mc css(chem.cas="34256-82-1")
# Should produce error:
calc mc css(chem.name="34256-82-1")
#Capitalization shouldn't matter:
calc mc css(chem.name="acetochlor"
calc_mc_css(chem.name="Acetochlor")
# What's going on?
help(calc mc css)
# What chemicals can I do?
get cheminfo()
```

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Oral Equivalent Dose Examples

#State-state oral equivalent dose (mg/kg BW/day) to produce 0.1 uM serum concentration for human, 0.95 quantile, for Acetochlor (published value):

get_wetmore_oral_equiv(0.1,chem.cas="34256-82-1")

#State-state oral equivalent dose (mg/kg BW/day) to produce 0.1 uM serum concentration for human, 0.95 quantile, for Acetochlor (calculated value):

calc_mc_oral_equiv(0.1,chem.cas="34256-82-1")

#State-state oral equivalent dose (mg/kg BW/day) to produce 0.1 uM serum concentration for human, 0.05, 0.5, and 0.95 quantile, for Acetochlor (published values):

get_wetmore_oral_equiv(0.1,chem.cas="34256-82-1",which.quantile=c(0.05,0.5,0.95))

#State-state oral equivalent dose (mg/kg BW/day) to produce 0.1 uM serum concentration for human, 0.05, 0.5, and 0.95 quantiles, for Acetochlor (calculated value):

calc_mc_oral_equiv(0.1,chem.cas="34256-82-1",which.quantile=c(0.05,0.5,0.95))

#State-state oral equivalent dose (mg/kg BW/day) to produce 0.1 uM serum concentration for rat, 0.95 quantile, for Acetochlor (calculated value):

calc_mc_oral_equiv(0.1,chem.cas="34256-82-1",species="Rat")

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Interspecies Extrapolation Examples

#Steady-state concentration (uM) for 1 mg/kg/day for 0.95 quantile for human for Acetochlor (calculated value): calc mc css(chem.cas="34256-82-1",method="dr"))

#Steady-state concentration (uM) for 1 mg/kg/day for 0.95 quantile for rat for Acetochlor (should produce errors since there is no published value, 0.5 quantile only):

get_wetmore_css(chem.cas="34256-82-1",species="Rat")

#Steady-state concentration (uM) for 1 mg/kg/day for 0.95 quantile for rat for Acetochlor (calculated value): calc mc css(chem.cas="34256-82-1",species="Rat")

#Steady-state concentration (uM) for 1 mg/kg/day for 0.5 quantile for rat for Acetochlor (published value): get_wetmore_css(chem.cas="34256-82-1",species="Rat",which.quantile=0.5)

#Steady-state concentration (uM) for 1 mg/kg/day for 0.5 quantile for rat for Acetochlor (calculated value): calc_mc_css(chem.cas="34256-82-1",species="Rat",which.quantile=0.5)

#Steady-state concentration (uM) for 1 mg/kg/day for 0.95 quantile for mouse for Acetochlor (should produce error since there is no published value, human and rat only):

get_wetmore_css(chem.cas="34256-82-1",species="Mouse")

#Steady-state concentration (uM) for 1 mg/kg/day for 0.95 quantile for mouse for Acetochlor (calculated value): calc_mc_css(chem.cas="34256-82-1",species ="Mouse")

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Help Files

Every function has a help file

help(add chemtable)

Add a table of chemical information for use in making httk predictions.

Description

This function adds chemical-specific information to the table chem.physical_and_invitro.data. This table is queried by the model parameterization functions when attempting to parameterize a model, so adding sufficient data to this table allows additional chemicals to be modeled.

Usage

add_chemtable(new.table, data.list, current.table=NULL, reference=NULL,species=NULL, overwrite=F)

Arguments

new.table Object of class data.frame containing one row per chemical, with each chemical minimally by described by a CAS

number.

data.list This list identifies which properties are to be read from the table. Each item in the list should point to a column in

the table new.table. Valid names in the list are: 'Compound', 'CAS', 'DSSTox.GSID' 'SMILES.desalt', 'Reference', 'Species', 'MW', 'logP', 'pKa Donor', 'pKa Accept', 'logMA', 'Clint', 'Clint.pValue', 'Funbound.plasma', 'Fgutabs',

'Rblood2plasma'. Note that Rblood2plasma (Ratio blood to plasma) is currently not used.

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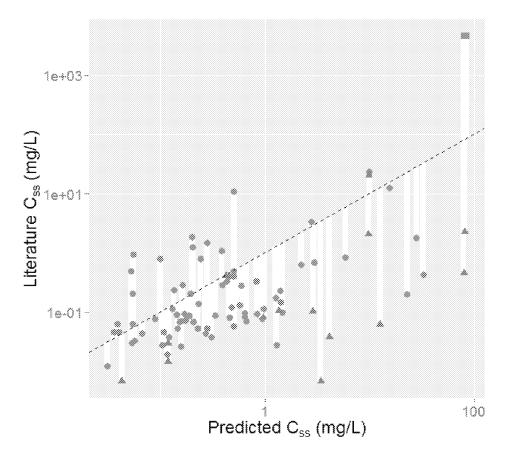
Why Do Statistical Analysis?

- In vivo Predictive Ability and Domain of Applicability
- In drug development, HTTK methods estimate therapeutic doses for clinical studies – predicted concentrations are typically on the order of values measured in clinical trials (Wang, 2010)
- For environmental compounds, there will be no clinical trials
- Uncertainty must be well characterized ideally with rigorous statistical methodology
 - We will use direct comparison to in vivo data in order to get an empirical estimate of our uncertainty
 - Any approximations, omissions, or mistakes should work to increase the estimated uncertainty when evaluated systematically across chemicals



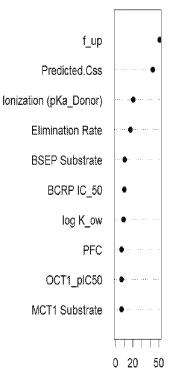
Using in vivo Data to Evaluate

RTK



Class * Pharmaceutical (74) * Other (11) * PFC (2)

- When we compare the C_{ss} predicted from *in* vitro HTTK with *in vivo* C_{ss} values determined from the literature we find limited correlation ($R^2 \sim 0.34$)
- The dashed line indicates the identity (perfect predictor) line:
 - Over-predict for 65
 - Under-predict for22
 - The white lines indicate the discrepancy between measured and predicted values (the residual)



Importance of Descriptors

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Wambaugh et al. (2015)



Predicting When RTK Will Work

- We can use computer algorithms to analyze chemical descriptors to try to predict when the residual will be small
- Factors included are:
 - Physico-chemical properties
 - Log(Kow), molecular weight, acid/base association constants (pKa), general pharmaceutical or perfluorinated compound classification
 - In vitro HTTK data
 - Plasma protein binding (F_{up}) and hepatic clearance
 - Active chemical transport
 - Use quantitative structure activity relationships (QSARs) to predict likelihood each compound is a substrate for 17 different transporters (From Alexander Sedykh and Alex Tropsha (UNC) and Sieto Bosgra (TNO))



Predicting RTK Errors

- The higher the C_{ss}, the lower the oral equivalent dose
- Ideally the residuals (difference between the literature value and the prediction) are small or $R \equiv C_{ss}^{lit.}/C_{ss}^{pred.} \approx 1$
- If a residual is large, we would prefer to over-predict C_{ss} to be conservative, *i.e.* R < 1

$$R = 1.02$$
 $F_{up} = 0.06$

$$R = 10$$
 $F_{up} = 0.08$

 $F_{up} < 0.11$

$$R = 0.5$$
 $F_{up} = 0.04$

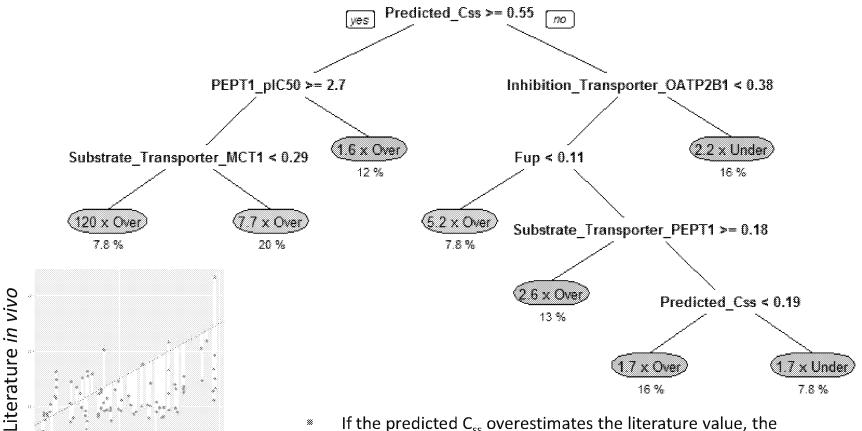
Nr.

YES R = 5 $F_{up} = 0.5$ $F_{up} = 0.08$

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Predicting HTTK Errors



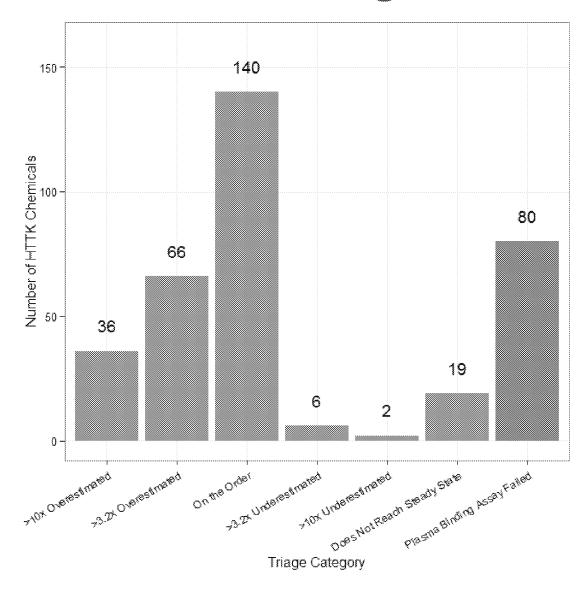
- Predicted from in vitro
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- If the predicted C_{ss} overestimates the literature value, the necessary exposure (i.e., equivalent dose) predicted with RTK will be lower
 - This is a conservative error for reverse dosimetry
- Worry about cases where we significantly underestimate necessary exposure

Wambaugh et al. (2015)



- Through comparison to in vivo data, a crossvalidated (random forest) predictor of success or failure of HTTK has been constructed
- Add categories for chemicals that do not reach steady-state or for which plasma binding assay fails
- All chemicals can be placed into one of seven confidence categories

Toxicokinetic Triage

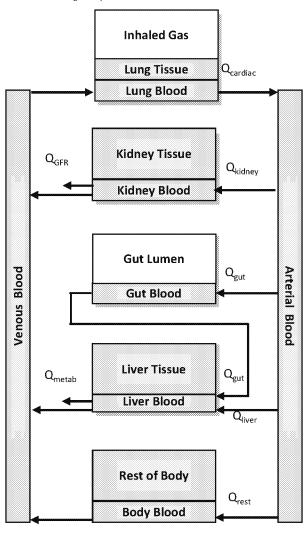


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Wambaugh et al. (2015)



A General Physiologically-based Toxicokinetic (PBTK) Model



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- "httk" also includes a generic PBTK model
- Some tissues (e.g. arterial blood) are simple compartments, while others (e.g. kidney) are compound compartments consisting of separate blood and tissue sections with constant partitioning (i.e., tissue specific partition coefficients)
- Exposures are absorbed from reservoirs (gut lumen)
- Some specific tissues (lung, kidney, gut, and liver) are modeled explicitly, others (e.g. fat, brain, bones) are lumped into the "Rest of Body" compartment.
- Blood flows move the chemical throughout the body.
 The total blood flow to all tissues equals the cardiac output.
- The only ways chemicals "leaves" the body are through metabolism (change into a metabolite) in the liver or excretion by glomerular filtration into the proximal tubules of the kidney (which filter into the lumen of the kidney).



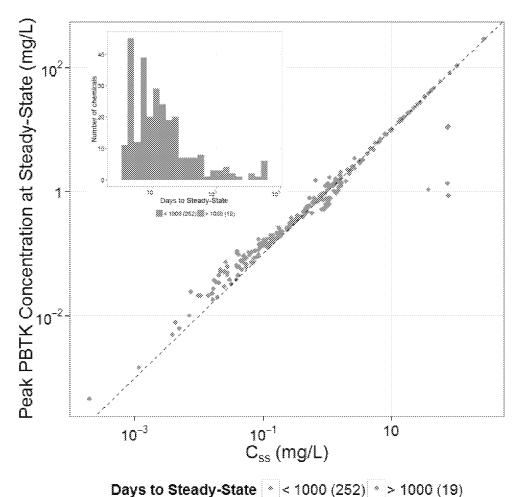
Basic PK Statistics Examples

```
library(httk)
#A Function to get PK summary statistics from the PBPK model:
help(calc stats)
# 28 day human study (20 mg/kg/day) for Abamectin:
calc stats(days=28,chem.name="bisphenol a", dose=20)
     Human plasma concentrations returned in uM units.
     AUC is area under plasma concentration curve in uM * days units with Rblood2plasma = 0.79.
     SAUC
     [1] 44.82138
     $peak
     [1] 23.16455
     $mean
     [1] 1.600764
# Units default to µM but can use mg/L:
calc stats(days=28,chem.name="bisphenol a", dose=20,output.units="mg/L")
# Same study in a mouse:
```

calc stats(days=28,chem.name="bisphenol a", dose=20,species="mouse")



Peak Concentration vs. C_{ss}



- Peak serum
 concentrations from the
 HTPBTK model are
 compared against the
 steady-state
 concentration predicted
 by the three
 compartment model for
 a constant infusion
 exposure (as in Wetmore
 et al. 2012)
- The dashed, identity (1:1) line indicates that for most compounds the peak concentrations are very similar to C_{ss}

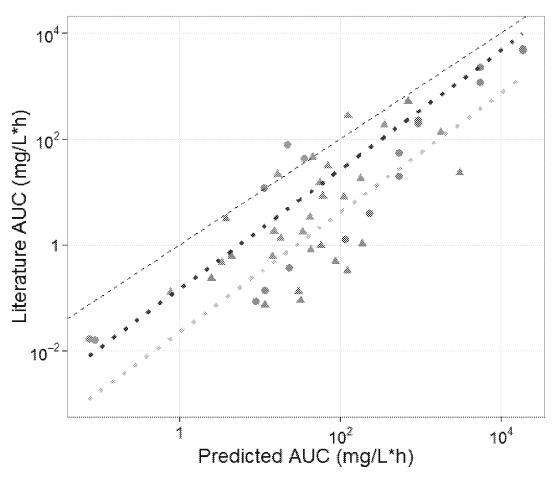
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Wambaugh et al. (2015)



Evaluating In Vitro PBTK Predictions with In Vivo Data



- PBTK predictions for the AUC (time integrated plasma concentration or Area Under the Curve)
- in vivo measurements from the literature for various treatments (dose and route) of rat.
- Predictions are generally conservative i.e.,
 predicted AUC higher than measured
 - Oral dose AUC ~6.4x
 higher than intravenous
 dose AUC

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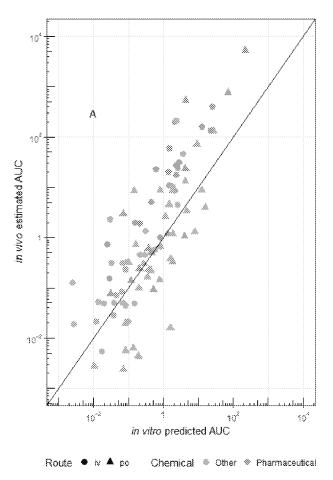
Route ᢀ iv ᢀ po ᢀ sc

Class ● Other (7) ▲ Pharmaceutical (15)

Wambaugh et al. (2015)



Analyzing New In Vivo Data (Rat)

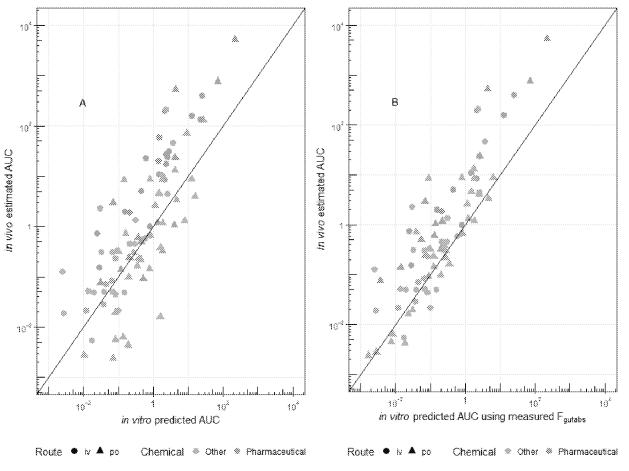


- Oral and *iv* studies for 26 ToxCast compounds
 - Collaboration with NHEERL (Mike Hughes and Jane Ellen Simmons)
 - Additional work by Research Triangle Institute (Tim Fennell)
- Can estimate
 - Fraction absorbed
 - Absorption Rate
 - Elimination Rate
 - Volume of Distribution

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Analyzing New In Vivo Data (Rat)



- Oral and *iv* studies for 26 ToxCast compounds
 - Collaboration with NHEERL (Mike Hughes and Jane Ellen Simmons)
 - Additional work by Research Triangle Institute (Tim Fennell)
- Can estimate
 - Fraction absorbed
 - Absorption Rate
 - Elimination Rate
 - Volume of Distribution

Cyprotex (ToxCast) is now measuring bioavailability (CACO2) for many HTTK chemicals

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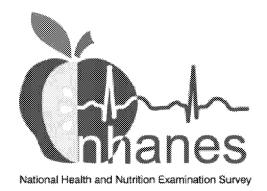


Population simulator for HTTK

Correlated Monte Carlo sampling of physiological model parameters

- Body weight
- Tissue masses
- Tissue blood flows
- GFR (kidney)
- Hepatocellularity

Source of data: CDC NHANES



Large, ongoing CDC survey of US population: demographic, body

measures, medical exam,

biomonitoring (health and exposure), ...

Designed to be representative of US population according to census data

Data sets <u>publicly available</u> (http://www.cdc.gov/nchs/nhanes.htm)

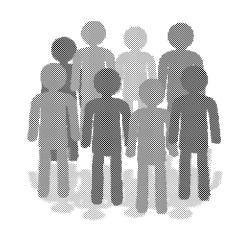
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Population simulator for HTTK

Sample
NHANES
quantities

Sex
Race/ethnicity
Age
Height
Weight
Serum creatinine



Regression equations from literature (+ residual marginal variability)

Predict
physiological
quantities

Tissue masses
Tissue blood flows
GFR (kidney
function)
Hepatocellularity

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(Similar approach used in SimCYP [Jamei et al. 2009], GastroPlus,
PopGen [McNally et al. 2014], P3M [Price et al. 2003], physB [Bosgra et al. 2012], etc.) Ring et al. (in press)



Generating demographic subgroups

User can specify	Default if not specified
Age limits	0-79 years
Sex (# males, # females)	NHANES proportions
Race/ethnicity (5 NHANES categories)	NHANES proportions
BMI/weight categories	NHANES proportions

- NHANES quantities sampled from appropriate conditional distribution (given specifications)
 - Physiological parameters predicted accordingly

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NHANES Demographic Examples

library(httk)

```
# Oral equivalent (mg/kg/day) for in vitro activity of 1 μM for Acetochlor calc_mc_oral_equiv(1,chem.cas="34256-82-1",method="dr")
```

```
# Oral equivalent (mg/kg/day) for NHANES "Mexican American" Population calc_mc_oral_equiv(1,chem.cas="34256-82-1",method="dr", reths = "Mexican American")
```

Oral equivalent (mg/kg/day) for NHANES "Mexican American" Population aged 18-25 years calc_mc_oral_equiv(1,chem.cas="34256-82-1",method="dr",agelim_years=c(18,25),reths = "Mexican American")

Probably too few individuals in NHANES for direct resampling ("dr") so use virtual individuals ("vi") resampling method:

calc_mc_oral_equiv(1,chem.cas="34256-82-1",method="vi",agelim_years=c(18,25),reths = "Mexican American")

Can also specify gender, weight categories, and kidney function

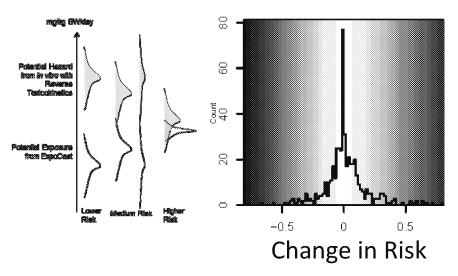
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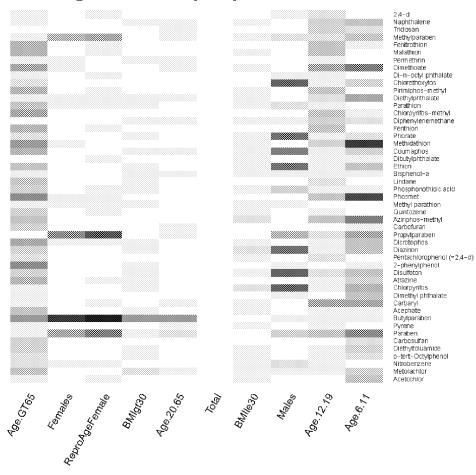


Life-stage and Demographic Specific Predictions

- Wambaugh *et al.* (2014) predictions of exposure rate (mg/kg/day) for various demographic groups
- Can use HTTK to calculate margin between bioactivity and exposure for specific populations



Change in Activity: Exposure Ratio





Version history for "httk"

The publicly available R package contains code and data that has been part of peer-reviewed publications (Old versions are archived)

- Version 1.1 accompanied "Toxicokinetic Triage for Environmental Chemicals" Wambaugh et al. (2015) Tox. Sci.
- Version 1.2 accompanied "httk: R Package for High-Throughput Toxicokinetics" Pearce et al., Journal of Statistical Software (*in press*)
- Version 1.3 accompanied "Incorporating High-Throughput Exposure Predictions with Dosimetry-Adjusted *In Vitro* Bioactivity to Inform Chemical Toxicity Testing" Wetmore et al., (2015) Tox. Sci.
- Version 1.4 addressed comments for acceptance of Pearce et al. (in press, J. Stat. Soft.)
- Version 1.5 accompanied "Identifying populations sensitive to environmental chemicals by simulating toxicokinetic variability," Ring et al. (in press, Env. International)
- Version 1.6 accompanied "Evaluation and Calibration of High-Throughput Predictions of Chemical Distribution to Tissues," Pearce et al. (submitted)
- Subsequent version numbers will be assigned as papers are accepted on:
 - Gestational model (Kapraun)
 - Inhalation exposure (Evans and Pearce)
 - New human data from Cyprotex (Wambaugh and Wetmore)
 - New rat data and revised IVIVE model (Honda)
 - More flexible PBPK model (Pearce)

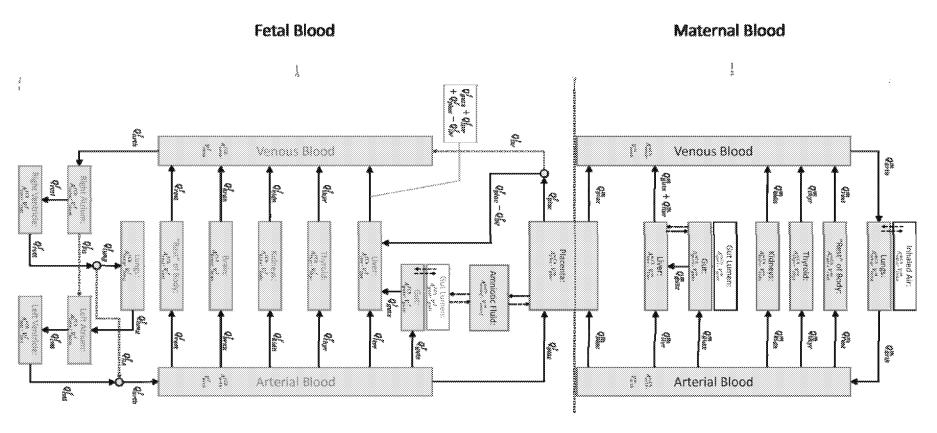
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Lead programmer Robert Pearce



Gestational Version of PBTK model Under Development



New httk package model that allows fetal tissue concentration predictions for all PBTK chemcials

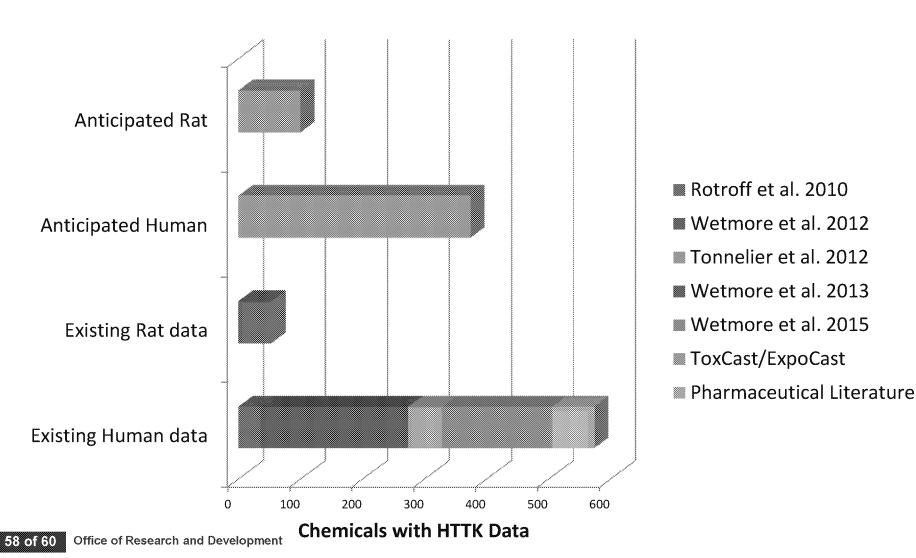
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Kapraun et al., (in preparation)



Chemicals with HTTK Data





Does My Chemical Have HTTK Data?

Is a chemical available?

> "80-05-7" %in% get_cheminfo()
[1] TRUE

> library(httk)

> get_cheminfo()

[1] "2971-36-0" "94-75-7" "94-82-6" "90-43-7" "1007-28-9"

[6] "71751-41-2" "30560-19-1" "135410-20-7" "34256-82-1" "50594-66-6"

[11] "15972-60-8" "116-06-3" "834-12-8" "33089-61-1" "101-05-3"

[16] "1912-24-9" "86-50-0" "131860-33-8" "22781-23-3" "1861-40-1" ...

> get_cheminfo(info="all")

All data on chemicals A, B, C

subset(get_cheminfo(info
="all"),Compound%in%c(
"A","B","C"))

		Human.Fu										
		pKa_Accep				Human.Cli	Human.Cli	nbound.pl	DSSTox_	DSSTox_Su Structure_		
Compoun	d CAS	logP	t I	oKa_Donor MV	V	nt	nt.pValue	asma	bstance_	_Id Formula	Substance	_Type
									DTXSID0	02 C8H6Cl2O		
2,4-d	94-75-7		2.81 <na></na>	2.81	221.03	3 (0.149	9 0.04	4 0442	3	Single	Compound
									DTXSID7	02 C10H10Cl2		
2,4-db	94-82-6		3.53 < NA>	4.5	249.09) (0.10	4 0.03	1 4035	O3	Single	Compound
2-												
phenylphe	e								DTXSID2	02		
nol	90-43-7		3.09 < NA>	10.6	170.211	L 2.08	0.16	4 0.04	4 1151	C12H10O	Single	Compound
6-												
desisoprop				DTXSID003								
ylatrazine	1007-28-9)	1.15 1.59	<na></na>	173.6	5 C	0.539	9 0.46	6 7495	C5H8ClN5	Single	Compoun

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Summary

- Toxicokinetics (TK) provides a bridge between HTS and HTE by predicting tissue concentrations due to exposure
- High Throughput (HTTK) methods developed for pharmaceuticals have been adapted to environmental testing
- A primary application of HTTK is "Reverse Dosimetry" or RTK
 - Can infer daily doses that produce plasma concentrations equivalent to the bioactive concentrations,
 - But: We must consider "domain of applicability"
- New R package "httk" freely available on CRAN allows statistical analyses to identify strengths and weaknesses
 - All HTTK models and data made public upon peer-reviewed publication

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Chemical Safety for Sustainability (CSS) Rapid Exposure and Dosimetry (RED) Project

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Antony Williams

NRMRL Yirui Liang Xiaoyu Liu NHEERL Linda Adams Christopher Ecklund Marina Evans Mike Hughes Jane Ellen Simmons

NERL Craig Barber Namdi Brandon Peter Egeghy Jarod Grossman Hongtai Huang Brandall Ingle Kristin Isaacs Sarah LaughlinToth Seth Newton Katherine Phillips

Paul Price
Jeanette Reyes
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John Streicher
Mark Strynar
Mike Tornero-Velez
Elin Ulrich
Dan Vallero
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Collaborators Arnot Research and Consulting

Arnot Research and Consulting Jon Arnot Battelle Memorial Institute Anne Louise Summer Anne Gregg **Chemical Computing Group** Rocky Goldsmith National Institute for Environmental Health Sciences (NIEHS) National Toxicology Program Mike Devito Steve Ferguson Nisha Sipes Netherlands Organisation for Applied Scientifi Research (TNO) Sieto Bosera Research Triangle Institute Timodov samal ScitoVation Harvey Cleviel Charte Micolas Silent Spring Institute Relate Decision Southwest Research Institute Alice Yau Kristin Favela Summit Toxicology Lesa Avlivard Tox Strategies Caroline Ring University of California, Davis Deporan Bermet Hyeone Med Shiri University of Michigan Olivier Jolliet

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Alex Troosha

Lead CSS Matrix Interface: John Kenneke (NERL)

The views expressed in this presentation are those of the authors and do not necessarily reflect the views or policies of the U.S. EPA



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